# Dear CAV, We Need to Talk About Reproducibility

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**Abstract** How many times have you tried to re-implement a past CAV tool paper, and failed?

Reliably reproducing published scientific discoveries has been acknowledged as a barrier to scientific progress for some time but there remains only a small subset of software available to support the specific needs of the research community (i.e. beyond generic tools such as source code repositories). In this paper we propose an infrastructure for enabling reproducibility in our community, by automating the build, unit testing and benchmarking of research software.

#### 1 Introduction

This is not a theory paper or a tool paper, nor an industrial case study. The primary aim of this paper is to start a discussion in the CAV community about the reproducibility of algorithms, models, tools and benchmarks across the computer-aided formal analysis and verification research domain. There is a significant opportunity for the CAV community to identify and address the technical and socio-cultural issues surrounding reproducibility in both the cognate research domain as well as more broadly for computer science; a desirable outcome would be a clear specification to encourage, enable and enforce reproducibility. Enumerating a standard for reproducibility would have a clear benefit for researchers as well as the CAV community as a whole.

We recognise that we do not need to sell the idea of reproducibility to the CAV community too much: we are used to writing papers about our algorithms, models and tools; and reproducing others' work (or having our own work reproduced) is encapsulated in the "Benchmark Tables" that CAV authors and referees both think are essential to any paper. But the idea of reproducibility is gaining momentum in the wider scientific community as well, for example in computer science [1], life sciences [2], psychology [3] and the social sciences [4]. There has been a revolution in the sharing and dissemination of published papers (open access) and the subsequent discussions relating to the sharing of protocols

and materials (open science) [5]. But the ability of a researcher to take published results and data and reimplement the described workflow remains difficult [6,7,8]. We have previously documented some of the technical and cultural barriers to reproducing work across computing and the computational sciences, both in terms of the sharing of algorithms [9] and models and benchmark sets [10,11].

A number of high-profile computer science conferences, including PLDI, POPL, SIGMOD, CGO, SPLASH and ECAI, now explicitly acknowledge the importance of reproducibility (and repeatability, recomputability and the multitude of 'Rs' that underpin e-research), as well as promoting community-driven reviewing and validation [12]. For many — the first time for CAV this year — this takes the form of the author providing artefacts (an accessible tool for reproducing results) to evaluate. Journals such as Nature, PLoS Computational Biology and Bioinformatics explicitly require that source code and data is made available online under some form of open source license. While these initiative are great, they are often optional, seem piecemeal, and do little to enable verification or validation of scientific results at a later stage. Even within the same field, there are different ideas of what defines reproducibility.

This paper is thus a "Call to Action", inviting CAV practitioners to embrace a new methodology for disseminating research. Practically, we propose an initial specification for a reproducibility service for CAV. We present the requirements of the prototype, and a suggested plan for introducing the tool to the community. Finally, we highlight key implementation issues relating to security and general applicability which will need mitigating or resolving before widespread acceptance by the research community. The benefits for reproducibility testing for a service are clear. In a sense, reproducibility here is an extension of the standard practice of testing. Three features distinguish our aims for reproducibility here: compilation and testing in a new machine, a continuous integration strategy for code commits, and the ability to add and remove benchmarks to the test set.

We have a running example based upon the BioModelAnalyzer (BMA)<sup>4</sup> tool, which has been described by the authors over a series of VMCAI and CAV papers [13,14,15]. BMA is a tool for the development and analysis (simulation, model checking) of a specific class of formal models for biology. The tool specifically allows users to test for model *stability*; that is, a bespoke algorithm proves that for all initial states a model always ends in a single fixpoint. We have chosen this example due to our familiarity with the tool, and to highlight historical examples where a reproducibility service would have supported both toolchain development and algorithm discovery.

# 2 A specification for reproducible computational science

A service for reproducibility is intended to play three important roles. It should:

1. Demonstrate that a piece of code can be compiled, run and behaves as described, without manual intervention from the developer.

<sup>4</sup> http://biomodelanalyzer.research.microsoft.com/

- Store and link specific artefacts with their linked publications or other publiclyaccessible datasets.
- 3. Allow new benchmarks to be added, by users other than the developer, to widen the testing and identify potential bugs.

## 2.1 de Novo build environments

A service such as this must require minimal developer intervention. This serves multiple purposes – through automation for example, the service can be enabled to compile new code and test new benchmarks trivially. This also forces the developer to make publicly available their local workarounds (i.e. hacks). As such, this requires the developer to make the project dependencies clearly available, and enables future changes in the dependencies (such as a library update) to be tested automatically too.

Throughout the lifetime of BMA, development has been shared between a number of developers, working on different aspects of the tool. Work in algorithm development focuses on adding new features to a command line tool with few dependencies, aiding rapid development. In contrast, the graphical model construction and testing environment has typically been done by a single or pair of individuals. This necessarily required a number of dependencies, reflecting the use of Azure (Microsoft's cloud computing platform) and Silverlight (a framework for rich Internet applications).

In an early stage of development it was found that only a single machine was capable of deploying the web service. This arose as the developer responsible for writing and deploying the user interface had run a series of commands necessary to run the mixture of 64- and 32- bit components on Azure. These commands needed only be run once, and went undocumented, thus needing to be rediscovered later when other team members attempted to deploy. These problems would be identified trivially through the proposed service; such undocumented commands would lead to all tests failing until explicitly added to the build process.

#### 2.2 There are two types of people

The service must fit easily into the developers workflow. As noted in Section 3 we expect that there will be some costs to the users in terms of the time required to ensure that the code compiles and runs on the service. To minimise this, the service needs to connect to standard code repositories, automatically detecting and responding to new versions of the code and updates to dependencies, running tests for every new code commit.

To address these needs we propose that the service should follow the workflow given in Figure 1. Two classes of user are defined; developers, who generate code; and modellers, who generate new benchmarks. An individual might in practice play either or both of these roles; here the roles serve to define ways in which people interact with the system. A developer writes new code, which is periodically pushed to a repository such as GitHub. Through integration with

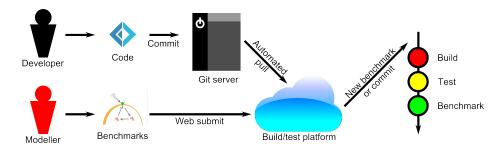


Figure 1. Proposed reproducibility service workflow

the repository, the server responds to new code by undergoing a process of pulling the code from the repository, downloading required dependencies, and compiling the code. If this stage fails, the developer is informed and the workflow ends. If the code is successfully compiled, two stages of testing are performed. The first stage (labelled *Test* in Figure 1), involves running a series of basic tests defined by the developer. This is intended as a sanity check to ensure that basic features of the code have not been broken by the updated code, and failure to pass these tests is reported and ends the workflow. If this completes successfully, the second stage (labelled *Benchmark* in Figure 1) a series of models are tested for a known property, and the results recorded. These results can then be stored in a database, with a note of the commit ID, and available through a web interface for future analysis.

### 2.3 Tool refinement

In contrast to the developer role, a modeller supplies benchmarks for a piece of code to test. These do not require that the latest version of the code is recompiled, but on submission the models are tested and added to the local repository of models for analysis.

In the case of BMA, throughout the development of the tool, many refinements have been made to different implementations. Some of these were subtle, and were identified by unit tests; for example rounding mechanisms were switched between floors and rounds following a scientific discussion. More complex changes however broke behaviours which were not tested in our available benchmark set. One example was in the treatments of nodes without inputs; "biological intuition" suggested that such nodes should have an alternative default function from other nodes. Here, the ability of users to submit new benchmarks would aid identification of these breaking changes, by extending the test sets and simplifying the process of adding to the test sets, and forcing the question of what changes are appropriate (and how to update old models to keep correct behaviour).

#### 2.4 Identification of algorithmic weaknesses

After a model is submitted, it is tested on every new piece of code pushed to the server and the changes in the behaviour can be noted and linked to specific code commits. Whilst the developer's role has a transparent value (in providing an implementation of an algorithm), the value of the modeller may be less immediately clear. The modeller submits a broad range of tests which may highlight material flaws (i.e. bugs) in the implementation, or the algorithm. More than this however, the modeller may generate models which identify weaknesses of either an algorithm or an implementation.

One example from the authors experience is the series of models with "timed-switches" described in [15]. There, we presented a new algorithm for proving stability in a new class of models. Whilst the paper focused on discussing the algorithm, identifying the new class of models was complex. Models with long cycles and the new class of models ("non-trivially stable models") both can take substantial time to search for cycles, and these models could only be proved stable using a combination of simulation (to identify the fix-point) and LTL queries (to prove that there existed no paths beyond a certain length which did not include the fix-point) [16].

#### 2.5 Algorithm-model axes

The proposed service would allow both types of test to be included explicitly, and models to be routinely added to each algorithm. Models which time-out with one but are successfully proved can be logged and identified for future study. The features which define them would be more easily found, and new algorithms developed to address the specific features of the model. It could further be used to demonstrate the speed improvement arising from new algorithms.

#### 2.6 makedepend

Dependencies for a given implementation need explicit testing. Due to the highly variable and sometimes complex nature of dependencies, we see this as an optional part of the workflow, as developers may chose to supply dependencies as binary files in the code compilation process. For completeness however we note that such a system could also respond to updates in external dependencies by triggering compilation and testing in the same manner as defined for a new code commit. This would aid developers in identifying code breaking changes introduced by third parties.

#### 2.7 "I'm first!"

Another issue is around performance comparisons of benchmarks: how can we estimate and compare raw performance in the cloud? Testing new algorithms on benchmarks is first about pass/fail, but very soon it's about raw performance. Benchmark tables are about beating other algorithms, other tools. But, if the

whole verification workflow is running on the cloud, then acquiring raw performance numbers is not possible any more. There is no cloud equivalent of top or time that gives user – resource statistics. There is too much infrastructure interference — with VMs spinning up, being torn down, migrating, the bus being used by other VMs, etc — to obtain faithful numbers for the user – process – VM itself. Projects such as Recomputation.org<sup>5</sup> have been focusing on using virtual machines in the cloud to freeze, and later unfreeze, computational experiments; while this approach is not the whole solution, it is certainly a move in the right direction [17]. Nevertheless, the project's primary aim is to validate recomputation, with performance a secondary consideration [18]. This is one aspect of this proposal that needs further investigating.

#### 2.8 Running arbitrary code

There are clearly security concerns around providing open e-infrastructure that pulls, compiles and runs arbitrary code as an autonomous continuous integration framework; we need to consider precisely how this infrastructure would interact with other open services, as well as privileges it would require to run as an autonomous cloud service.

# 3 A reproducibility model for the CAV community

Following the proposal of such a system, the question becomes: how do we encourage widespread uptake, or even standardisation? Such a service may appear non-trivial, given the large numbers of tools and workflows that could potentially require to be supported by the service. Furthermore, after such a service has been implemented, how do we ensure it is useful and usable for researchers. To address this, we propose the following workflow for CAV:

- 1. **Pre-conference:** clear signposting for authors; it should be advertised and promoted in the CAV call for papers to highlight this is a step-change in how we address reproducibility. Call for artefact reviewers with a range of specialisms, with a named chair of the review team.
- 2. Explicit criteria for authors: make this as easy as possible for us to evaluate/execute your artefact!. We would aim to articulate the review criteria, but the primary aim is: can I evaluate/execute this artefact and get the same results that are presented in the paper?
- 3. **Submission:** when papers are submitted, they have to nominate whether they want their paper to go through artefact review (at the start, this may not be compulsory, but this will change over a period of time effecting cultural change and this would then become a necessary condition and a formal stage in the reviewing workflow), along with required tools, libraries and (ideally) computational requirements.

<sup>5</sup> http://recomputation.org/

- 4. **Reviewing:** in the first instance, it may be seen as an extra (voluntary) step to the normal reviewing process: e.g. *This submission is voluntary and will not influence the final decision regarding the papers.*. Independent of the scientific merit of the paper, the results will be verified. To encourage this, there may be a prize, as well as ranked ordering and profiled listed in conference proceedings.
- 5. **Artefact evaluation:** artefact evaluation process runs concurrent to the standard paper review process.
- 6. **Reporting:** traffic lights system (potentially with ranked list) to indicate the level of reproducibility of the submitted artefact.
- 7. **Community curation:** over a number of CAV cycles, we would have a community curated repository/database of previous artefacts, which would provide exemplars, comparisons and emerging best practice.

The key question for different research communities then becomes: how to initialise this change? Such a requirement creates a set of new costs to researchers, both in terms of time spent ensuring that their tools work on the centralised system (in addition to their local implementation), but also potentially in terms of equipment (in terms of running the system). Such costs may be easier to bear for some groups compared to others, especially those with large research groups who can more easily distribute the tasks, and it is important that the service does not present a barrier to early career researchers and those with efficient budgets (this type of cost analysis is not unique to reproducibility efforts – it has been estimated that a shift to becoming exclusively open access for a journal may lead to a ten-fold increase in computer science publication costs [19]).

#### 4 Conclusions

The benefits to the community from a cultural change to favour reproducibility are clear and as such we should aim through the software infrastructure and the CAV workflow to mitigate these costs. Furthermore, we can reasonably expect the needs of the community to evolve over time, and initial implementations of the platform may require refinement in response to user feedback. As such, if the community is to move to requiring reproducibility, it seems most reasonable that this is staggered over a number of years to allow for both of these elements to develop, until eventually all authors are required to use the service. This plan balances competing needs within the community, and would reduce the disruption for uptake by gradually introducing it to researchers.

- CAV 2016: Offer the service as an optional extra in the testing phase, allowing users to demonstrate the reliability of their code which could be taken into account in the review process.
- CAV 2017: All authors must use the reproducibility service, but results are not used in the review process. The results of the test are used to refine the service and pick out any unaddressed issues

**CAV 2018:** All authors are required to use the service, and the results are explicitly used to assess reproducibility in the review process.

An open discussion and understanding of what reproducibility means for the CAV community is important: we need to explicitly state that this is worthwhile and address it, or don't bother doing it at all. We thus propose that this should be trialled for CAV in 2016, with an explicitly-defined three year schedule for universal adoption by CAV 2018.

#### References

- Collbery, C., Proebsting, T., Moraila, G., Shankaran, A., Shi, Z., Warren, A.M.: Measuring Reproducibility in Computer Systems Research. Technical report, Department of Computer Science, University of Arizona (2014)
- Rollins, N.D., Barton, C.M., Bergin, S., Janssen, M.A., Lee, A.: A Computational Model Library for publishing model documentation and code. Environmental Modelling & Software 61 (2014) 59

  –64
- Chambers, C.D., Feredoes, E., Muthukumaraswamy, S.D., Etchells, P.J.: Instead
  of "playing the game" it is time to change the rules: Registered Reports at AIMS
  Neuroscience and beyond. AIMS Neuroscience 1(1) (2014) 4–17
- Conte, R. et al.: Manifesto of Computational Social Science. The European Physical Journal Special Topics 214(1) (2012) 325–346
- 5. Royal Society: Science as an open enterprise. (2012) Available from: https://royalsociety.org/policy/projects/science-public-enterprise/report/.
- Peng, R.D.: Reproducible Research in Computational Science. Science 334(6060) (2011) 1226–1227
- Sandve, G., Nekrutenko, A., Taylor, J., Hovig, E.: Ten Simple Rules for Reproducible Computational Research. PLoS Computational Biology 9(10) (2013)
- Wilson, G. et al.: Best Practices for Scientific Computing. PLoS Biology 12(1) (2014)
- 9. Crick, T., Hall, B.A., Ishtiaq, S.: "Can I Implement Your Algorithm?": A Model for Reproducible Research Software. In: Proceedings of 2nd International Workshop on Sustainable Software for Science: Practice and Experiences (WSSSPE2). (2014)
- Crick, T., Hall, B.A., Ishtiaq, S., Takeda, K.: "Share and Enjoy": Publishing Useful and Usable Scientific Models. In: Proceedings of 1st International Workshop on Recomputability. (2014)
- 11. Crick, T., Chue Hong, N.P., eds.: Proceedings of the 1st International Workshop on Recomputability, IEEE Press (2014) http://recomputability.researchcomputing.org.uk/.
- 12. Fursin, G., Dubach, C.: Community-Driven Reviewing and Validation of Publications. In: Proceedings of the 1st ACM SIGPLAN Workshop on Reproducible Research Methodologies and New Publication Models in Computer Engineering (TRUST'14), ACM Press (2014) 1–4
- 13. Cook, B., Fisher, J., Krepska, E., Piterman, N.: Proving stabilization of biological systems. In: Proceedings of the 12th International Conference on Verification, Model Checking and Abstract Interpretation (VMCAI'11). (2011)
- Benque, D., Bourton, S., Cockerton, C., Cook, B., Fisher, J., Ishtiaq, S., Piterman, N., Taylor, A., Vardi, M.Y.: BMA: Visual Tool for Modeling and Analyzing Biological Networks. In: Proceedings of the 24th International Conference on Computer Aided Verification (CAV 2012). (2012)

- 15. Cook, B., Fisher, J., Hall, B.A., Ishtiaq, S., Juniwal, G., Piterman, N.: Finding instability in biological models. In: Proceedings of the 26th International Conference on Computer Aided Verification (CAV 2014). (2014)
- Claessen, K., Fisher, J., Ishtiaq, S., Piterman, N., Wang, Q.: Model-Checking Signal Transduction Networks through Decreasing Reachability Sets. In: Proceedings of the 25th International Conference on Computer Aided Verification (CAV 2013). (2013)
- 17. Arabas, S. et al.: Case Studies and Challenges in Reproducibility in the Computational Sciences. Available from: http://arxiv.org/abs/1408.2123 (April 2014)
- Gent, I.P.: The Recomputation Manifesto. Available from: http://arxiv.org/ abs/1304.3674 (April 2013)
- 19. Vardi, M.Y.: Openism, IPism, Fundamentalism, and Pragmatism. Communications of the ACM 57(8) (2014)